

NATIONAL RESEARCH UNIVERSITY



DO MERGERS POLICIES INCREASE UNIVERSITIES' EFFICIENCY? EVIDENCE FROM REGRESSION DISCONTINUITY DESIGN

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OUTLINE

- Universities mergers in Russia
- Policy description
- Literature on mergers in HE sector
- Quasi-experimental evaluation of mergers' impact on university efficiency
- Discussion and concluding remarks

MERGERS IN RUSSIAN HE



Romanenko and Lisytkin, 2018

MONITORING OF PERFORMANCE

Introduced in 2012 by Ministry of Science and Higher Education

-Gather the data on performance indicator and define threshold values

- -If particular university fails to overcame thresholds for at least 4 indicators out of 7, it receives "inefficient" from the Ministry
- -"inefficient" university may be merged to stronger university

MERGER IS CONSIDERED IN CASE LESS THAN 4 PERFORMANCE INDICATORS ARE MET

Monitoring of performance as a basis for merger policy



LITERATURE ON MERGERS IN HE

Motivations for mergers activities

Rowley, 1997 Botha, 2001 Fazackerley, 2017 Short-run and long-run effects of merger policies on different sides of universities' activities

Valimaa et al., 2014 Wan, 2008 Factors affecting merger process

Harman, 2002; Locke, 2007 Kyvik and Stensaker, 2013 Case-studies of universities mergers in different countries

Aagaard et al., 2016 Harman and Meek, 2002 Harman and Harman, 2003

MOTIVATIONS FOR MERGERS

Strategic mergers

-Mergers in order to improve positions in international ratings (Valimaa et al., 2014)

-Mergers in order to achieve national public policy aims (Wang, 2001)

-Mergers in order to expand growth capabilities (Johnes and Tsionas, 2018)

Economically motivated mergers

-Mergers in order to improve performance (Fielden and Markham, 1997)

-Mergers in order to reduce costs (Johnes and Tsionas, 2018)

-Mergers in order to improve efficiency (Johnes, 2018)

LITERATURE ON MERGERS IN HE

Mergers and efficiency of higher education institutions

China:

Hu and Liang, 2008; Mao et al., 2009: efficiency gains can be observed only in the first year after the merger

UK:

<u>Johnes, 2014</u>: Average efficiency is significantly higher among merged than either pre-merger or non-merging universities

<u>Papadimitriou and Johnes, 2018</u>: Merged universities on average demonstrate higher efficiency when controlling for observed heterogeneity – subject mix, source of income, size. The strongest effect of merger on efficiency – in the first year after merger

<u>Johnes and Tsionas, 2018</u>: inefficiency is negatively affected by tendency to merge and the action of merging. High heterogeneity of effect of merger on efficiency across different cases of mergers.

RESEARCH QUESTION

Do university mergers affect their efficiency (performance to resources ration)?

Possible channels:

1) Economies of scale effect:

-spread of administrative costs over larger output
-spread of faculty over larger number of students
-spread of fixed assets maintenance cost over larger output

2) Economies of scope effect:

-merger between universities with different missions (more efficient production of teaching and research jointly)

-development of interdisciplinary research after merger of universities with different specializations

3) Quality of management

METHODOLOGY STEP 1: PSM SAMPLE

Propensity score matching (PSM) is a technique allows homogenizing the sample and building relevant control group

Initial data:

447 public universities without branches

38 universities with "inefficient status" in 2013

14 universities that were consequently merged

Final data:

152 universities (matching 3:1)

PSM SAMPLE

		Before PSM	
	Treated	Non-treated	Whole sample
Average USE score (e1)	62.33 (6.50)	65.06 (7.95)	64.79 (7.85)
Total amount of R&D	172	210	206
projects per faculty (e2)	(178)	(361)	(347)
Share of foreign students (e3)	3.34 (3.47)	4.36 (5.16)	4.26 (5.02)
Total income from all sources per faculty (e4)	1850 (714)	2048 (1330)	2029 (1284)
Total area of training and laboratory facilities per student (e5)	12.68 (4.14)	15.86 (8.63)	15.54 (8.35)
Employment of graduates (e6)	96.75 (3.5)	98.13 (1.86)	97.99 (2.12)
Faculty with PhD per 100 students (e7)	4.27 (1.44)	15.33 (23.73)	14.24 (22.77)

PSM SAMPLE

		After PSM	
	Treated	Control	Whole PSM sample
Average USE score	62.33	62.62	62.55
	(6.50)	(5.94)	(6.06)
Total amount of R&D projects per faculty	172	185	181
	(178)	(473)	(419)
Share of foreign students	3.34	2.79	2.93
	(3.47)	(3.11)	(3.2)
Total income from all	1850	1856	1854
sources per faculty	(714)	(918)	(869)
Total area of training and laboratory facilities per student	12.68 (4.14)	12.82 (4.64)	12.79 (4.51)
Employment of graduates	96.75	97.22	97.10
	(3.5)	(2.48)	(2.76)
Faculty with PhD per 100 students	4.27 (1.44)	4.31 (1.31)	4.30 (1.34)

STEP 2: EFFICIENCY ESTIMATION

Bootstrapped DEA Estimator (Simar and Wilson, 2000)

INPUTS



- ✓ Total income from all sources x_1
- ✓ Total number of academic staff x_2
- ✓ Average entrance exam score x_3

Robustness checks:

-SFA translog, SFA Cobb-Douglas, QR

-Additional inputs and outputs: total square of buildings used for teaching and research activities, share of faculty with advanced degrees, total volume of private R&D projects.

OUTPUTS

- ✓ Number of publications in index journals per 100 of academic staff y_1
- \checkmark Total amount of R&D projects (in rubles) y_2
- \checkmark Total number of employed graduates- y_3



EFFICIENCY ESTIMATION

	2013 (Before		2017 (After treatment)			Change 2013-2017, %			
	tr	eatment	z)						
	Treated	Non- treated	Whole sample	Treated	Non- treated	Whole sample	Treate d	Non- treated	Whole sample
				Inpu	its				
Total income									
from all sources, ml	3 232 (2 140)	1 067 (1 132)	1 266 (1 397)	2 816 (1 847)	1 046 (1 242)	1 209 (1 399)	87.2	98.0	95.5
roubles									
Average USE score	(7.5)	65.5 (9.0)	66.0 (9.0)	71.9 (8.4)	64.5 (9.0)	65.2 (9.2)	101.1	98.5	98.8
Total number of faculty	1208 (502)	518 (387)	581 (445)	1 001 (359)	425 (357)	478.8 (393.4)	82.9	82.3	82.4
~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~~				Outp	uts				
Total number of publications	· 2 537 (2 215)	522 (559)	707 (1024)	4 193 (3 158)	1 733 (1 862)	1 960 (2 124)	165.3	331.9	277.2
Total number of graduates	· 12 908 (4 670)	5 452 (3 895)	6139 (4508)	12 804 (5 065)	4 928 (4 384)	5 653 (4 627)	99.2	90.4	92.1
Total amount									
of R&D projects, ml. roubles	487 (488)	108 (198)	142 (261)	657 (679)	129 (273)	178 (362)	134.7	120.0	124.5

# **EFFICIENCY ESTIMATION**

	Treated	Non-treated	Whole sample
2013	0.829	0.777	0.781
	(0.054)	(0.105)	(0.102)
2017	0.797	0.714	0.722
	(0.064)	(0.125)	(0.121)
Change 2013-2017, %	96.1	91.9	92.5

density.default(x = eff_2013)



# **METHODOLOGY STEP 3: FRDD**

 $Y_i = \beta_0 + \beta_1 T_i + \beta_2 f(PS_i) + \beta X + \epsilon_i$ 

 $T_i = \alpha_0 + \alpha_1 T A_i + \alpha_2 h(PS_i) + \alpha X + \vartheta_i$ 

 $Y_i$  – an efficiency change of university *i* between 2013 and 2017;

 $T_i$  – a dummy variable representing that university was merged;

 $PS_i$  – a performance score of university *i* in 2013 (average ratio of observed value to threshold value);

f(.), h(.) – flexible functional forms;

X – the matrix of control variables (total number of students, average USE score, share of students in STEM fields);

 $\epsilon_i$ ,  $\vartheta_i$  – random errors;

 $TA_i$  – a dummy variable reflecting university was assigned to the treatment;

## **METHODOLOGY STEP 3: FRDD**



We regress outcome variable (efficiency gain) on the assignment variable (performance score). If we observe a gap in the regression line on the threshold, we observe the effect of the policy

## **STEP 3: FRDD**

#### The structure of sample





#### Descriptive statistics for control variables

	2013 (Before		2017 (After treatment)			Change 2013-2017, %			
	treatment)								
	Tractad	Non- Whole		Tracted	Non-	Whole	Treated	Non-	Whole
	Treated	treated	sample	Treated	treated	sample		treated	sample
Average USE score	72.1 (7.0)	65.5 (9.0)	<b>66.</b> 1 (9.0)	73.1 (7.7)	64.5 (9.0)	65.3 (9.2)	101.4	98.5	98.8
Total number of students	13 179 (4 745)	5 455 (3 882)	6 115 (4 501)	13 147 (5 101)	4 922 (3 970)	5 626 (4 669)	99.8	90.2	92.0
Share of students in STEM fields	0.5 (0.3)	0.4 (0.3)	0.4 (0.3)	0.5 (0.3)	0.4 (0.3)	0.4 (0.3)	-	-	-

# **FRDD: RESULTS**

	(1)	(2)	(3)	(4)	(5)	(6)	(7)		
Parametric estimation (2SLS)									
Treatment effect	0.072	0.118	0.1142	0.157**	0.184*	0.137*	0.154*		
(standard error)	(0.063)	(0.074)	(0.075)	(0.059)	(0.071)	(0.065)	(0.075)		
Polynomial of performance score	First	First	First	Second	Second	Third	Third		
Controls	No	Yes	Yes	No	Yes	No	Yes		
Interactions	No	No	Yes	No	Yes	No	Yes		
# of observations	152	152	152	152	152	152	152		
	Non-p	arametric	estimation	(local line	ar)				
Treatment effect	0.271*	0.313*	0.324*	0.275*	0.287*	0.173*	0.244*		
(standard error)	(0.106)	(0.124)	(0.146)	(0.113)	(0.141)	(0.081)	(0.127)		
Bandwidth	Optimal (0.195)	Optimal (0.195)	Optimal (0.195)	Optimal * 0.5 (0.097)	Optimal * 0.5 (0.097)	Optimal * 2 (0.391)	Optimal * 2 (0.391)		
Controls	No	Yes	Yes	No	Yes	Yes	Yes		
Interactions	No	No	Yes	No	Yes	Yes	Yes		
# of observations	83	83	83	56	56	104	104		

## **FRDD: RESULTS**





Third order polynomial

Performance score

# DISCUSSION

- Merged universities experienced greater efficiency gains (smaller efficiency declines) after the merger was implemented
- However, this effect can be identified just ner a cutoff point
- Possible channels through which merger process may influence efficiency: economies of scale effect, economies of scope effect, changes in managerial practices.
- The treatment effect identified using regression discontinuity design can be interpreted as a total influence of all possible factors.

# **EFFICIENCY ESTIMATION**



# **STEP 3: FRDD – THRESHOLD IDENTIFICATION**

Literature on structural breaks: (Card et al., 2008; Steinberg, 2014)

 $ineff_i = \beta_0 + \beta_1(Threshold_i) + \epsilon_i$ 

where *inef f_i* is a variable representing the "inefficient status" received by university *i* from the Ministry based on the Monitoring of Performance in 2013; *Threshold_i* is an indicator function of the form:

 $Threshold_i = I\{APS_i < \theta\}$ 

where  $APS_i$  is an aggregate performance score for university *i*;  $\theta$  is the threshold value that we have to identify from our sample; *I* is an indicator function.

# **STEP 3: FRDD – THRESHOLD IDENTIFICATION**

to achieve greater robustness we consider four alternative specifications:

 $ineff_{i} = \beta_{0} + \beta_{1}(Threshold_{i}) + \beta_{2}(APS_{i}) + \epsilon_{i}$   $ineff_{i} = \beta_{0} + \beta_{1}(Threshold_{i}) + \beta_{2}(APS_{i}) + \gamma X_{i} + \epsilon_{i}$   $ineff_{i} = \beta_{0} + \beta_{1}(Threshold_{i}) + f(APS_{i}) + \epsilon_{i}$   $ineff_{i} = \beta_{0} + \beta_{1}(Threshold_{i}) + f(APS_{i}) + \gamma X_{i} + \epsilon_{i}$ 

where  $\gamma X_i$  is the set of university's characteristics multiplied by the set of regression coefficients: total number of students on the university, average entrance exam score, share of students in STEM field; f(.) is a second-order polynomial function; all other notations remain the same.

# **STEP 3: FRDD – THRESHOLD IDENTIFICATION**

 $Threshold_i = I\{APS_i < \theta\}$ 

Each equation we estimated 40 times with different values of  $\theta$ ,  $\theta \in [0.8; 1.2]$ 

The highest R-squared corresponds to the  $\theta$ =1.07 (in all equations)

Therefore, 1.07 is identified as a potential discontinuity point